

EE558 - Digital Communications

Lecture 2: Review of Signals and Systems

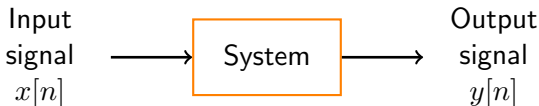
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Leadership Starts Here

Signals and Systems



■ Signal

- ▶ Applied to something that conveys information
- ▶ Represented as a function of one or more independent variables
- ▶ Continuous-time vs. Discrete-time
- ▶ Continuous-amplitude vs. Discrete-amplitude

- System: A transformation or operator that maps a input sequence into an output sequence

$$y[n] = T(x[n]) \quad \text{or} \quad y(t) = T(x(n)).$$

Signals

- Discrete-time signal $x[n]$

$$E_{\infty} = \sum_{n=-\infty}^{\infty} |x[n]|^2, \quad P = \lim_{T \rightarrow \infty} \frac{1}{2T} \sum_{n=-T}^T |x[n]|^2 \quad (1)$$

- Some signals have infinite average power, energy or both
- A signal is called an **energy signal** if $E_{\infty} < \infty$
- A signal is called an **power signal** if $0 < P_{\infty} < \infty$
- A signal can be an energy signal, a power signal, or neither type
- A signal cannot be both an energy signal or a power signal
- Examples: $x[n] = 1$, $x[n] = \sin n$, $x[n] = n$

Some Examples

- Time shift: $x[n - n_0]$
- Time reversal: $x[-n]$
- Time scaling: $x[an]$
- Periodic signal with period N : $x[n] = x[t + N]$
- Even signal: $x[-n] = x[n]$
- Odd signal: $x[-n] = -x[n]$
- Exponential signal: $x[n] = Ce^{an}$
 - ▶ Real-valued exponential vs Complex exponential
 - ▶ Growing or decaying?
 - ▶ Periodic or aperiodic?
- Real sinusoidal signal: $x[n] = A \cos(\omega n + \phi)$

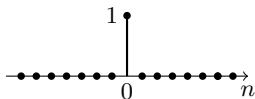
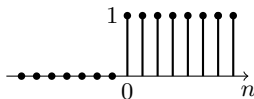
Unit Step Function and Unit Impulse

- Unit step function

$$u[n] = \begin{cases} 0, & n < 0 \\ 1, & n > 0 \end{cases}$$

- Unit impulse function

$$\delta[n] = u[n] - u[n - 1], \quad u[n] = \sum_{m=-\infty}^n \delta[m]$$



- Some properties:

- ▶ $\sum_{n=-\infty}^{\infty} x[n]\delta[n - n_0] = x[n_0]$: sifting property

- ▶ $x[n] = \sum_{k=-\infty}^{\infty} x[k]\delta[n - k]$: signal decomposition

Linearity

- Input-output relationship: $y_i[n] = T(x_i[n])$
- A system is linear if
 - ▶ $T(ax[n]) = aT(x[n])$
 - ▶ $T(x_1[n] + x_2[n]) = T(x_1[n]) + T(x_2[n])$
 - ▶ or $y[n] = T(a_1x_1[n] + a_2x_2[n]) = a_1y_1[n] + a_2y_2[n]$.
- Examples: linear or not
 - 1 Time scaler: $y[n] = x[2n]$
 - 2 Amplifier: $y[n] = 2x[n] + 1$
 - 3 Accumulator: $y[n] = \sum_{k=-\infty}^n x[k]$
 - 4 Squarer: $y[n] = x^2[n]$

Causality and Stability

- **Causality:** Output only depends on values of the input at only the present and past times

- Examples: casual or not

- ① Time scaler: $y[n] = x[2n]$ and $y[n] = x[n/2]$

- ② $y[n] = \sin(x[n])$

- **Stability:** Small input lead to responses that do diverge

$$|x[n]| \leq B \text{ for some } B < \infty \longrightarrow |y[n]| < \infty$$

- Examples: stable or not

- ① $y[n] = nx[n]$

- ② $y[n] = e^{x[n]}$

- ③ $y[n] = y[n - 1] + x[n]$

Time-Invariance

- Time-invariant system: characteristics of the system are fixed over time

$$y[n] = T(x[n]) \quad \longrightarrow \quad y[n - n_0] = T(x[n - n_0])$$

- Examples: Time-invariant or not

① $y[n] = \sin x[n]$

② $y[n] = nx[n]$

③ $y[n] = x[2n]$

- Linear time-invariant (LTI) system: good model for many real-life systems

- Examples: LTI or not

① $y[n] = \frac{1}{2n_0} \sum_{k=n-n_0}^{n+n_0} x[k]$

Response in LTI Systems



- Impulse response: Response to a unit impulse
- Any signal can be expressed as a sum of impulses

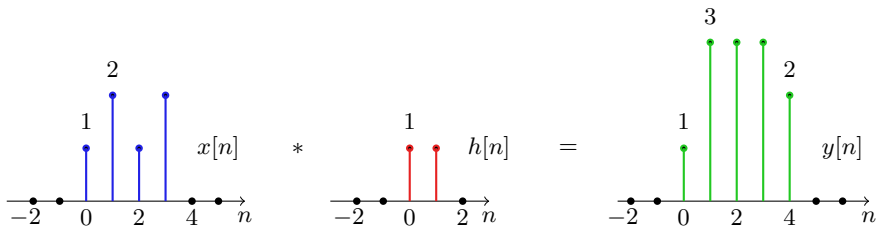
$$x[n] = \sum_{k=-\infty}^{\infty} x[k]\delta[n - k]$$

- LTI system: $\delta[n - k] \rightarrow h[n - k]$
- Output signal:

$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n - k]$$

Convolution Operation

- Convolution operation: $y[n] = x[n] * h[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$
- Commutative: $x[n] * h[n] = h[n] * x[n]$
- Associative: $x[n] * (h_1[n] * h_2[n]) = (x[n] * h_1[n]) * h_2[n]$
- Distributive: $x[n] * (h_1[n] + h_2[n]) = x[n] * h_1[n] + x[n] * h_2[n]$
- Examples: Flip, shift, multiply and add



LTI System Properties and Impulse Response

- Any LTI system can be described by its impulse response
- Memoryless: $h[n] = a\delta[n]$
- Causal: $h[n] = 0, \forall n < 0$
- Stable: $\sum_{n=-\infty}^{\infty} |h[n]| < \infty$

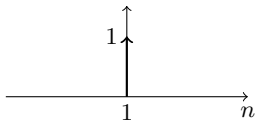
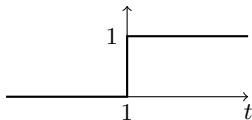
Continuous time Signals

- Unit step function

$$u(t) = \begin{cases} 0, & t < 0 \\ 1, & t > 0 \end{cases}$$

- Unit impulse function or Dirac delta function

$$\delta(t) = \frac{du(t)}{dt}, \quad u(t) = \int_{-\infty}^t \delta(\tau) d\tau$$



- $\delta(t) = 0$ for $t \neq 0$
- $\delta(t)$ is unbounded at $t = 0$
- $\int_{-\infty}^{\infty} \delta(t) dt = 1$ and $\int_{-\infty}^{\infty} x(t) \delta(t - t_0) dt = x(t_0)$: sifting property

Response in LTI Systems



- Impulse response: Response to a unit impulse
- Any continuous-time signal can be expressed as

$$x(t) = \int_{-\infty}^{\infty} x(\tau)\delta(t - \tau)d\tau$$

- LTI system: $\delta(t - \tau) \rightarrow h(t - \tau)$
- Output signal:

$$y(t) = \int_{-\infty}^{\infty} x(\tau)h(t - \tau)d\tau \triangleq x(t) * h(t)$$

- Examples: $x(t) = e^{-at}u(t)$, $h(t) = u(t)$. Then, $y(t) = \frac{1}{-a} [1 - e^{-at}]$.

Response to Complex Exponentials

- Input signal: $x(t) = e^{st}$

- Output signal:

$$y(t) = e^{st} \int_{-\infty}^{\infty} h(\tau) e^{-s\tau} d\tau = H(s) e^{st}$$

- $H(s)$ at s : eigenvalue associated with the eigenfunction e^{st}

- Input signal: $x[n] = z^n$

- Output signal:

$$y[n] = z^n \sum_{k=-\infty}^{\infty} h[k] z^{-k} = H(z) z^n$$

- $H(z)$ at z : eigenvalue associated with the eigenfunction z^n

- Why is eigenfunction is important?

- Can any signal be represented as a summation of complex exponentials?

Fourier Series I

- Periodic signal with period T : $x(t) = x(t + T)$
- $\omega_0 = 2\pi/T$ is called the “angular fundamental frequency”
- $f_0 = 1/T$ is called the “fundamental frequency”
- Harmonically related complex exponentials: $\Phi_k(t) = e^{jk\omega_0 t}$
- Assume a periodic signal $x(t)$ can be represented as

$$\text{Synthesis form : } x(t) = \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t}$$

- Coefficients a_k 's

$$\text{Analysis form: } a_k = \frac{1}{T} \int_T x(t) e^{-jk\omega_0 t} dt$$

Fourier Series II

- Fourier Analysis using fundamental frequency $f_0 = \omega_0/(2\pi)$
 - ▶ Synthesis form:

$$x(t) = \sum_{k=-\infty}^{\infty} a_k e^{jk2\pi f_0 t}$$

- ▶ Analysis form:

$$a_k = \frac{1}{T} \int_{-T/2}^{T/2} x(t) e^{-jk2\pi f_0 t} dt$$

- Parseval's theorem

$$\frac{1}{T} \int_{-T/2}^{T/2} x^2(t) dt = \sum_{k=-\infty}^{\infty} |c_k|^2$$

- Examples: A periodic square wave

Fourier Transform

- A periodic square wave & Fourier Coefficients

$$x(t) = \begin{cases} 1, & |t| < T_1 \\ 0, & T_1 < |t| < T/2 \end{cases}, \quad a_k = \frac{2 \sin(k\omega_0 T_1)}{k\omega_0 T}$$

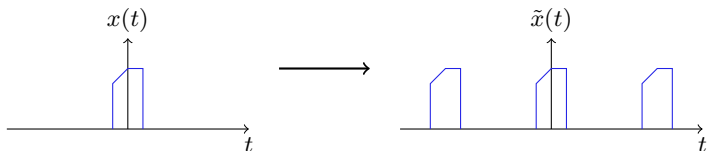
- Envelop function

$$T a_k = \left. \frac{2 \sin \omega T_1}{\omega} \right|_{\omega=k\omega_0}$$

- Fourier series coefficients and their envelop with different values of T with T_1 fixed
- $T \rightarrow \infty$: Fourier series coefficients approaches the envelope function.

Fourier Transform I

- Aperiodic signal: can be treated as a periodic signal with $T \rightarrow \infty$
- The envelop function is called the Fourier Transform
- Derivations of Fourier Transform
 - ▶ Period padding for a aperiodic signal $x(t)$ with finite duration



Fourier Transform II

- Express $\tilde{x}(t)$ using Fourier Series

$$\tilde{x}(t) = \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t}$$

where the Fourier Series coefficients are

$$a_k = \frac{1}{T} \int_T \tilde{x}(t) e^{jk\omega_0 t} dt = \frac{1}{T} \int_{-\infty}^{\infty} x(t) e^{jk\omega_0 t} dt$$

Define $X(j\omega) = \int_{-\infty}^{\infty} x(t) e^{-j\omega t} dt$: Analysis Equation of Fourier

Transform, then $a_k = \frac{1}{T} X(jk\omega_0)$. Thus,

$$\tilde{x}(t) = \sum_{k=-\infty}^{\infty} \frac{1}{T} X(jk\omega_0) e^{jk\omega_0 t} = \sum_{k=-\infty}^{\infty} \frac{1}{2\pi} X(jk\omega_0) e^{jk\omega_0 t} \omega_0$$

Fourier Transform III

- ▶ As $T \rightarrow \infty$, $\omega_0 \rightarrow 0$

$$\lim_{\omega_0 \rightarrow 0} \sum_{k=-\infty}^{\infty} \frac{1}{2\pi} X(jk\omega_0) e^{jk\omega_0 t} \omega_0 = \int_{-\infty}^{\infty} \frac{1}{2\pi} X(j\omega) e^{j\omega t} d\omega$$

As $\tilde{x}(t) \rightarrow x(t)$, Synthesis Equation of Fourier Transform of $x(t)$:

$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\omega) e^{j\omega t} d\omega$$

- Fourier Transform can be applied to periodic and aperiodic signals.
Fourier Series can only be applied to periodic signals
- Examples: $x(t) = e^{-at}u(t)$ for $a > 0$

Properties of Fourier Transform I

- Linearity: if $x_1(t) \longleftrightarrow X_1(j\omega)$ and $x_2(t) \longleftrightarrow X_2(j\omega)$

$$a_1x_1(t) + a_2x_2(t) \longleftrightarrow a_1X_1(j\omega) + a_2X_2(j\omega)$$

- Time shifting: $x(t - t_0) \longleftrightarrow e^{-j\omega t_0} X(j\omega)$

- Conjugate: $x^*(t) \longleftrightarrow X^*(-j\omega)$

- Differentiation and Integration:

$$\frac{d}{dt}x(t) \longleftrightarrow j\omega X(j\omega)$$

$$\int_{-\infty}^t x(\tau)d\tau \longleftrightarrow \frac{1}{j\omega} X(j\omega) + \pi X(0)\delta(\omega)$$

- Time scaling: $x(at) \longleftrightarrow \frac{1}{|a|} X\left(\frac{j\omega}{a}\right)$

Properties of Fourier Transform II

- Parseval Equality: $\int_{-\infty}^{\infty} |x(t)|^2 dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} |X(j\omega)|^2 d\omega$
- Duality: Suppose $x(t) \longleftrightarrow X(j\omega)$ and $y(t) \longleftrightarrow Y(j\omega)$. If $y(t)$ has the shape of $X(j\omega)$, then $Y(j\omega)$ has the shape of $x(t)$
Example: $\delta(t) \longleftrightarrow 1$
- Convolution: $x(t) * h(t) \longleftrightarrow X(j\omega)H(j\omega)$
- Multiplication: $x(t)h(t) \longleftrightarrow \frac{1}{2\pi} X(j\omega) * H(j\omega)$
- Fourier Transform can often be denoted as $X(f)$ instead of $X(j\omega)$

$$X(f) = \int_{-\infty}^{\infty} x(t)e^{-j2\pi ft} dt$$
$$x(t) = \int_{-\infty}^{\infty} X(f)e^{j2\pi ft} df$$

Frequency Transfer Function

- LTI system: $y(t) = x(t) * h(t) = \int_{-\infty}^{\infty} x(\tau)h(t - \tau)d\tau$

- Fourier transform: $Y(f) = X(f)H(f)$

- Fourier transform of the impulse response function

$$H(f) = \int_{-\infty}^{\infty} h(t)e^{-j2\pi ft} dt$$

is called *frequency transfer function* or the *frequency response*

- $H(f) = |H(f)|e^{j\theta(f)}$
 - ▶ $|H(f)|$: magnitude response
 - ▶ $\theta(f)$: phase response

- Examples: $x(t) = A \cos 2\pi f_0 t$, output will be

$$y(t) = A|H(f_0)| \cos [2\pi f_0 t + \theta(f_0)]$$

Distortionless Transmission

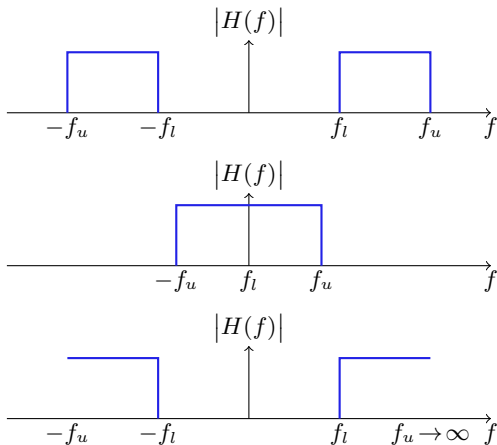
- Ideal system with constant delay and amplifier $y(t) = Kx(t - t_0)$
- Fourier Transform from both sides: $Y(f) = KX(f)e^{-j2\pi ft_0}$
- Transfer function

$$H(f) = Ke^{-j2\pi ft_0}$$

- Ideal distortionless transmission: constant magnitude response and its phase shift must be linear with frequency
- In practice, a signal will be distorted by some parts of a system
- Phase or amplitude correction (equalization) may be required for correction

Ideal Filter

- No ideal network exists: $|H(f)| = K, \forall f \rightarrow$ infinite bandwidth
- Truncated network: all frequencies in $[f_l, f_u]$ without distortion
- Passband: $f_l < f < f_u$, bandwidth $W_f = f_u - f_l$



Ideal Bandpass Filter

- Constant magnitude response

$$|H(f)| = \begin{cases} 1 & \text{for } |f| < f_u \\ 0 & \text{for } |f| \geq f_u \end{cases}$$

- Linear phase response: $e^{-j\theta(f)} = e^{-j2\pi f t_0}$

- Impulse response of the ideal low-pass filter

$$\begin{aligned} h(t) &= \mathcal{F}^{-1}\{H(f)\} = \int_{-\infty}^{\infty} H(f)e^{j2\pi f t} df \\ &= 2f_u \frac{\sin 2\pi f_u(t - t_0)}{2\pi f_u(t - t_0)} \end{aligned}$$

- What is wrong with this impulse response function?
- Realizable filters: Butterworth filter, Raised-cosine filter, etc